



Advanced Information Systems Technologies (AIST)

Machine Learning in NASA's Earth Science Division



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Outline of Talk

- Hope, Tragedy, and Myths
- What's it Going to Take
- Analytic Center as a Framework
- AIST Managed Cloud Environment (AMCE)
- Arc-GIS Online (AGOL)
- Lessons Learned
- Backup



Hope,

- A more robust understanding of a natural phenomenon or physical process
 - Relying on conclusions based on 3,000 examples, not 3
 - Digest massive volumes of observational data
 - High volume output from high-resolution ES models
 - Conventional techniques drown ES researchers
- Prediction of future state for complex systems from observational data without a thorough understanding of complex or poorly understood underlying physics
- Coordination of elements of a sensor web to target transient and transitional phenomena
 - Leverage emergence of SmallSats
 - Autonomy
 - Detection of interesting features and re-tasking to observe them
- Fast comparative OSSEs
- Assimilation models which do better at washing out the experimental error while preserving features and physical anomalies



Tragedy,

- Large investments by DoD and IC in AI and Machine Learning have, largely, not been leveraged by NASA Earth Science
- Lack of credibility in ML by Earth Science researchers
 - Why does ML generate bogus conclusions?
 - Failure to articulate assumptions & constraints and exceeding them
 - Overtraining
 - Drift over time
 - Failure to recognize when conditions change
 - Earth Scientists have trouble understanding the technology
 - Lack of demonstration of value to motivate them
 - Lack of training in the tools and algorithms
- Partnerships among Machine Learning and Earth Science communities are slow to take root
 - Communications
 - Opportunity
 - Trust
- Data grooming takes huge amounts of time



Myths

- Do a problem, publish for others to use, move on to next.
 - Reality: There is no fire and forget in Earth Science
- Open source is always better than commercial software
 - GIS is a good counter-example
 - Communicating with other Agencies requires arc-GIS capability
- Data must be supplied in same form as collected
 - Reality: Most investigators transform data once they get it
 - Often fail to verify measurements, without introducing errors
 - Must be able to trace back to authoritative data
- Cannot re-use (trust) anyone else's work
 - Community acceptance, peer review could create common capabilities
 - Journal publications want to help with this
 - Apache wants to help with this
 - Fix the data once and others could use it if they believe the fix



What's it Going to Take?

- **Community Acceptance** in NASA Earth Science (ES)
 - NASA Earth Scientists must understand the processes & algorithms
 - “In the field of observation, fortune favors the prepared mind.” – Pasteur
 - Collaboration among Earth Scientists, Computer Scientists and Technologists
 - Demonstration of value to the science communities
 - New Science
 - Faster Science
 - Results backed up by confirmation with legacy techniques
 - Uncertainty Quantification
 - Engagement with NASA ES by experienced ML practitioners
- **Easy to use** tools and credible results
 - To Earth Scientists
 - Which make available accepted processes and algorithms
- **Improved Data Usability**
 - Rapid use of the numeric results without significant grooming

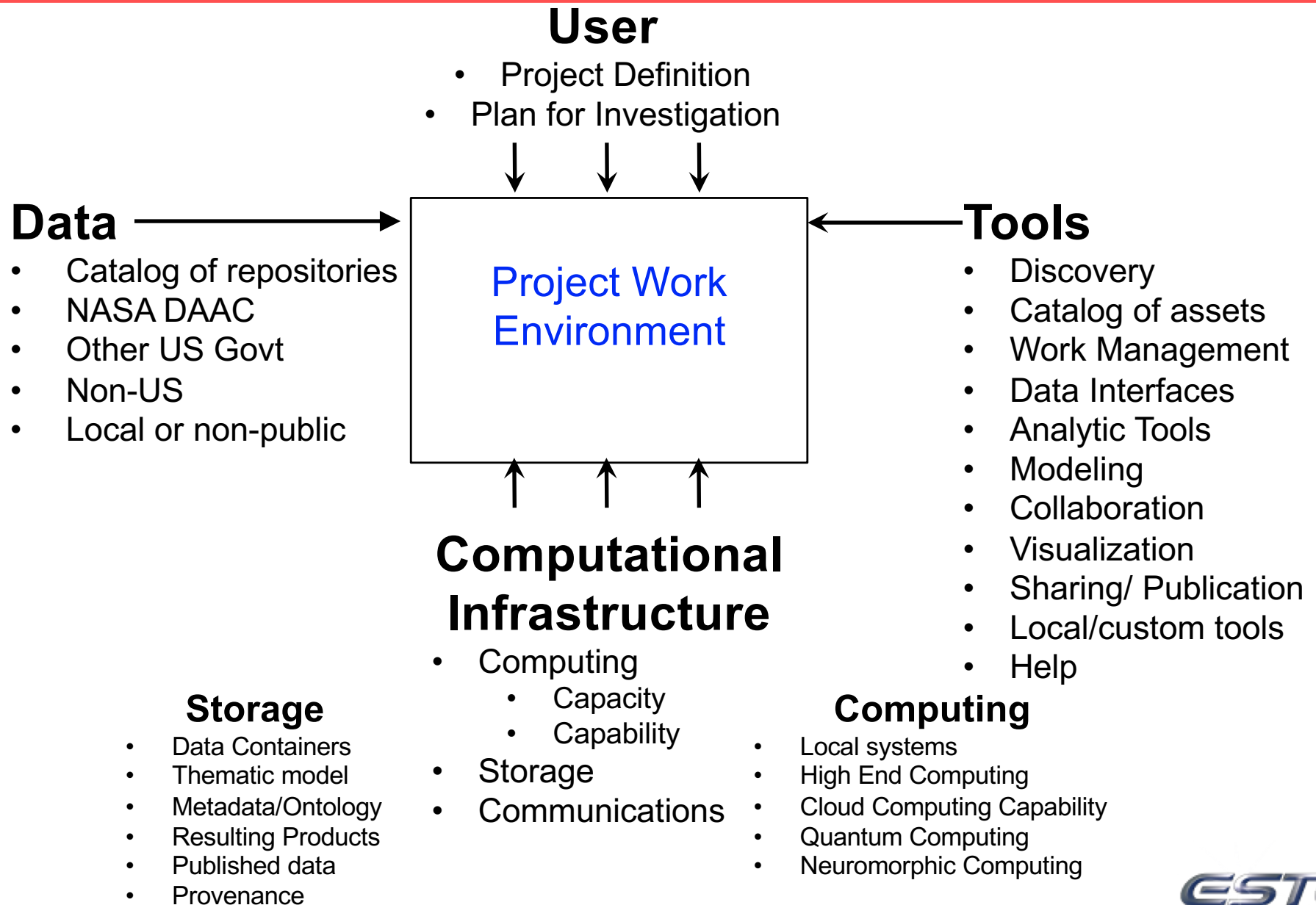


What is an Analytic Center

- An environment for conducting a Science investigation
 - Enables the confluence of resources for that investigation
 - Tailored to the individual study area
- Harmonizes data, tools and computational resources to permit the research community to focus on the investigation
 - Reduce the data preparation time to something tolerable
 - Catalog of optional resources (think HomeDepot shopping)
 - Semantic-enabled catalog of resources (think Yelp) with help
 - Relevant publications
 - Provide established training data sets of varying resolution
 - Provide effective project confidentiality, integrity and availability
 - Single sign-on and unified financial tracking

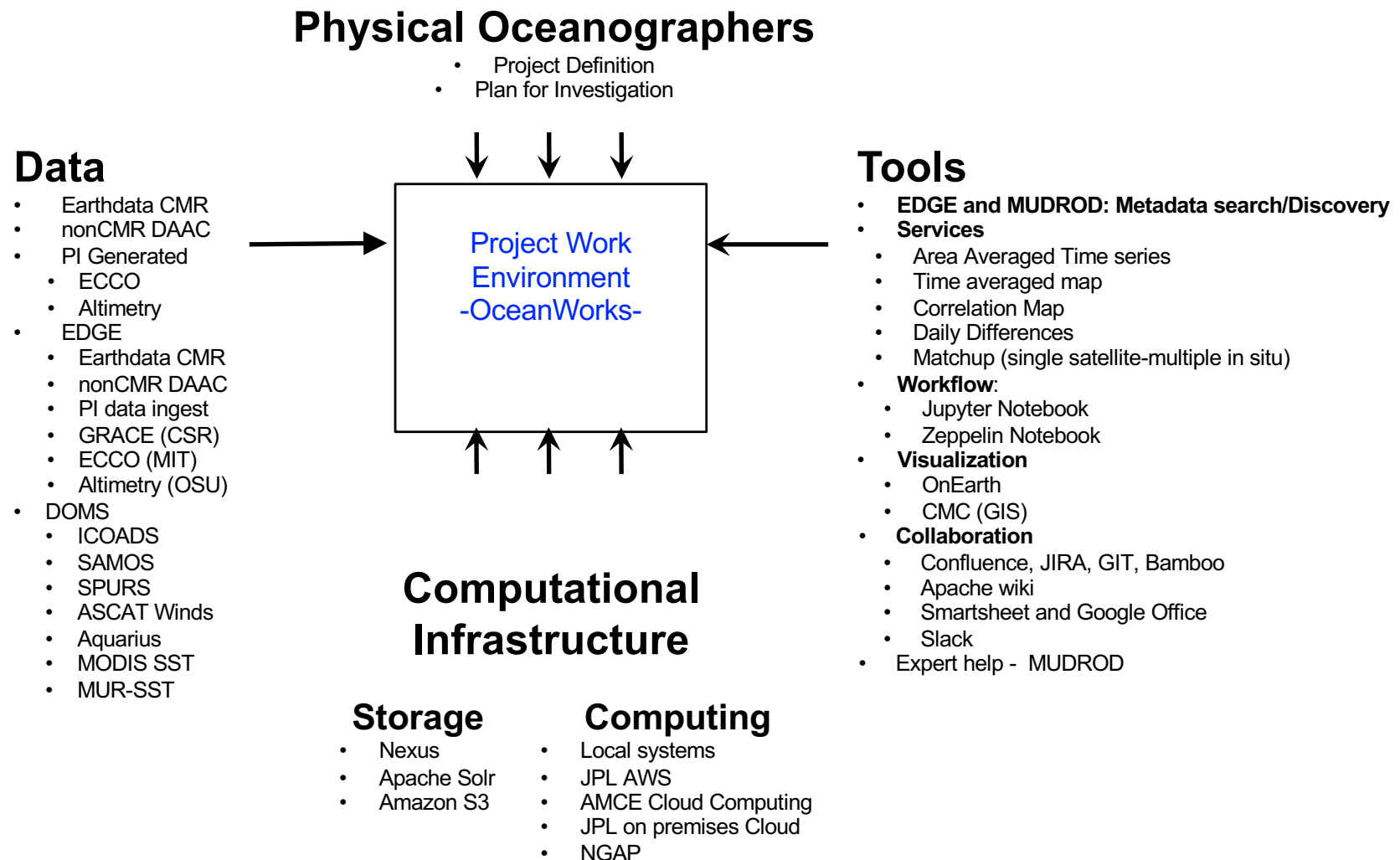


Analytic Center as a Framework





OceanWorks as an Example Analytic Center



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Analytic Center Characteristics

- Little to NO overhead when not being used
 - Persistent sources of funding to support infrastructure for science are hard to find
- Seamless integration of new/custom tools
- Comprehensive catalog of tools and known data (from many repositories)
 - Clear applicability (or not)
 - Expert system as an operator aid
- Help in using them
 - YouTube, moodle



Theory of Data Systems (ToDS)

- ToDS is SAMSI collaboration on how should data systems support analysis
 - NASA, JPL, NOAA, NSF collaborating to support ToDS
 - NASA Point of Contact: Amy Braverman at JPL
- Primary Focus is on **Uncertainty Quantification** in multi-source data analysis
- Most Science Investigations need data from **MULTIPLE SOURCES**
 - Each data supplier focuses on their own data
 - Different geolocation reference point in each data product
 - Different flags for no-observation
 - Quality flags instead of error but often no uncertainty quantification
 - Different grid references
 - Different statements/estimates of uncertainty and error
- NASA Mission design focuses on stewardship and Cal/Val
 - Format is unique to the instrument science team and the data product
- Large volumes of data from disparate sources require modern analytic techniques, tools and computing capacity
 - Regression techniques lack techniques to quantify uncertainty



Computing Resources

- Analysis of Earth Science phenomena requires large scale computing
 - Analysis of observations of many events (examples of phenomena or processes) requires large quantities of computing
 - Modeling often requires highly interconnected nodes to handle non-linear phenomena
 - Visualization requires appropriate numeric files to make meaning obvious
- Computing choices available
 - NASA High-end Computing Centers (NAS and NCCS)
 - Other Government supercomputing sites in conjunction with colleagues
 - Commercial Cloud Computing
 - AWS
 - Google
 - Microsoft
 - Suppliers of specialized cloud computing
- Use of computing requires funding and an Authorization to Operate (ATO)
 - Most investigations are in the LOW category
 - Limited NASA need for Moderate category in Science investigations
 - Many require collaboration with foreign nationals, which is hard to do inside the NASA address space



AMCE Management Considerations

- AIST demonstrated that Scientific Investigations can be conducted by NASA in a public cloud computing environment
 - Must support the way in which NASA ESD scientific investigations are conducted
- Analogous to developing an apartment building
 - Provide maximum flexibility for residents without impacting other residents
 - Not all projects belong in a multi-tenant facility
 - Some projects need to be in a single family dwelling
- Fundamental Principles
 - Provide Project PI with maximum flexibility in determining what to use in AWS
 - Ensure that projects are protected from each other
 - Help the PI ensure that his projects do not over run their budget
 - Automate financial overrun notices: alert, warning and shutdown
 - Monitor compliance with essential Computer Security restrictions
 - Automate notification of problems
- AIST assumes NASA Earth Science Data will be supplied via EOS-DIS efforts
 - AMCE provides a complementary service for PI's to buy computing near the data
 - AMCE is not intended to provide a repository for data
- NCCS is in process of operationalizing AMCE
 - Eventual expansion into Google and Microsoft Azure environments



PI's Receive Benefits of Cloud Computing

- Full Range of NASA Investigation Teams on a single computing environment
 - University, JPL, NASA Center, Other Government Agencies and Industry
 - Supercomputing requirements filled through normal HEC processes
- Startup Leadtime can be as little as 5 days
- Multi-party collaboration in neutral environment
 - NASA and non-NASA individuals and organizations
 - Exists outside NASA address space so no opportunity to contaminate NASA systems
- Use funding to buy time on AMCE instead of authorizing purchase of hardware
 - AIST directly funds transition and learning curve
 - Research Project funds consumption of AWS services after transition with 10% overhead
 - History indicates this cost is typically far lower (i.e., 10% or less) of investment cost
 - Other partners can buy their own AWS time and yet integrate into the computing environment
- Codes developed in AMCE are more easily integrated into NASA computing systems
- Similar needs elsewhere in Science Mission Directorate (SMD) Research divisions



Experiments with an ATO for arc-GIS Online (AGOL)

- AGOL is a cloud-based version of arc-GIS online allowing for collaboration with non-NASA data, users and colleagues
- Accepted challenge due to widespread requests for accounts on Esri Enterprise License Agreement (ELA)
 - Users were unable to obtain an ATO for their individual projects
 - FEDRAMP Process was stalled (although moving at this time)
- Opportunity to experiment with supplying data from arc-GIS Enterprise Servers to AGOL
- ATO Letter is expected to be ready for signature within two weeks (3/1/18)
 - ATO for Software-as-a-Service is unfamiliar ground at NASA
 - Approval expected at LOW Computer Security rating
 - ONLY Public Data



AIST Strategy

- Create an environment for conducting scientific investigations with data, tools and computing resources
 - Machine Learning workshop in April - ESIP
- Fund collaborations among scientific investigators, data experts and analytics experts
 - AIST-18 will fund experiments in the use of data analytics, not just technology development
 - Multi-disciplinary teams with the goal of enabling science
 - Draft solicitation in April, 2018 for community comment



AC Considerations in CA

- Focus must be on enabling individual scientific investigations
 - Maximum flexibility must be afforded to the PI
 - Accelerate the process of identifying, acquiring and cleaning up the data for their use
 - Enable access to both NASA and non-NASA data and tools
- Framework must make it easy to add new tools and capabilities without complex integration
 - Most investigators prefer the tools they are comfortable with and understand
 - Must be able to integrate contributions from others
- Framework must have very low overhead cost
 - No cost to keep it around when not being used
- Role of the DAACs in Analytic Centers
 - Provide effective, fast interfaces to the data products they hold
 - Allow scientists to extract the data of interest from archive
 - Provide assistance in the analysis and interpretation of their data products
 - Curation of well-documented Training Sets for Supervised Machine Learning
 - Support the science communities for their themes
 - Aid in the development of analytic tools related to their thematic responsibilities



Backup

- AIST
 - Responsibilities
 - AIST Projects with Machine Learning Components
- A Sample of Indian Projects using Machine Learning
- Machine Learning in China
- Machine Learning in Europe
- Useful Tools and Algorithms
- Analytic Center Needs





A Possible Framework for Adoption

- Focus on needs of a particular Science community
 - Data, models and tools they are comfortable with
 - Computational resources adequate to perform their investigation
 - Concept of Operations must support their work
- NASA's Earth Science needs **much** more exposure to Machine Learning
 - Researchers and Managers need to develop confidence in the tools
 - Outreach to the Science community
- Must be an architecture with modularity and interoperability
- Must support security
- Increase access to non-NASA data sources as well as NASA
- AIST has been experimenting with basic concepts as well as components



AIST Experiments with Analytic Centers

- Land Use and Land Change
 - NEX (NASA Ames) Rama Neimani
- Physical Oceanography
 - OceanWorks (JPL) Thomas Huang
- Tropical Cyclones and Hurricanes
 - TCIS (JPL) Svetla Hristova-Veleva
- Climate
 - Climate Model Diagnostic Analyzer (CMDA), Seungwon Lee (JPL)
 - Climate Workbench (UAH/MSFC) Manil Maskey, Chris Lynnes
- Communities being discussed for further experiments
 - Biodiversity
 - Hydrology
 - Cryosphere



AIST-14 Tasks-Machine Learning

| PI Name | Org | Prop # | Project Title | ESTO PM | End Date | Machine Learning Elements |
|----------------------|---------------|--------|---|---------|----------|--|
| Kamalika Das | UC Santa Cruz | 115 | Uncovering Effects of Climate Variables on Global Vegetation | Oza | 5/31/17 | Symbolic regression |
| Milton Halem | UMBC | 96 | Computational Technologies: Feasibility Studies of Quantum Enabled Annealing Algorithms for Estimating Terrestrial Carbon Fluxes from OCO-2 and the LIS Model | Cole | 5/31/17 | RBM, CNN, data assimilation |
| Hook Hua | JPL | 109 | Agile Big Data Analytics of High-Volume Geodetic Data Products for Improving Science and Hazard Response | Norton | 5/31/17 | Fault recognition and Science processing redirection |
| Thomas Huang | JPL | 28 | OceanXtremes: Oceanographic Data-Intensive Anomaly Detection and Analysis Portal | Norton | 8/29/17 | NEXUS as a data delivery tool - oceanography |
| Kristine Larson | U of CO | 4 | AMIGHO: Automated Metadata Ingest for GNSS Hydrology within OODT | Hines | 5/31/17 | Sensor characterization |
| Victor Pankratius | MIT | 36 | Computer-Aided Discovery of Earth Surface Deformation Phenomena | Little | 5/31/17 | Computer aided discovery |
| Chaowei Yang | GMU | 82 | Mining and Utilizing Dataset Relevancy from Oceanographic Dataset (MUDROD) Metadata, Usage Metrics, and User Feedback to Improve Data Discovery and Access | Cole | 5/31/17 | Natural language processing, CNN, SVM, deep learning |
| Tomasz Stepinski | Cincinnati | 27 | Pattern-based GIS for Understanding Content of very large Earth Science datasets | Quam | 6/31/17 | Classification and similarity |
| Jonathan Gleason | LaRC | 95 | Ontology-based Metadata Portal for Unified Semantics (OlyMPUS) | Oza | 1/31/17 | Precision ontology foundation |
| Constantine Lukashin | LaRC | 14 | NASA Information And Data System (NAIADS) for Earth Science Data Fusion and Analytics | Murray | 1/31/17 | Scientific enterprise service bus |
| Aashish Chaudhary | Kitware | 65 | Visualization Pipelines for big-data on the NASA Earth Exchange (NEX) Prototyping Agile Production, Analytics | Hines | 3/31/17 | Workflow |
| Martyn Clark | UCAR | 88 | Development of Computational Infrastructure to Support Hyper-resolution Large-ensemble Hydrology Simulations from Local-to-Continental Scales | Hines | 4/30/17 | Assimilation and ensembles |
| Kwo-Sen Kuo | Bayesics | 56 | DEREChOS: Data Environment for Rapid Exploration and Characterization of Organized Systems | Little | 4/30/17 | Foundation for data delivery |
| Seungwon Lee | JPL | 32 | Climate Model Diagnostic Analyzer | Norton | 4/30/17 | |
| Christian Mattman | JPL | 34 | SciSpark: Highly Interactive and Scalable Model Evaluation and Climate Metrics for Scientific Data and Analysis | Norton | 4/30/17 | Foundation for data delivery |



AIST QRS Tasks-Machine Learning

| PI Name | Org | Prop # | Project Title | ESTO PM | End Date | Machine Learning Elements |
|---------------|-----------|------------------|--|---------|----------|---|
| Dan Duffy | GSFC | AIST-QRS-16-0002 | Demo of VR Technology with live data(Phase 2) & GIS (Phase 3) | Cole | 12/31/16 | VR |
| Sreeja Nag | ARC | AIST-QRS-16-0003 | Scheduling Satellite Pointing within Constellations | Hines | 12/31/16 | |
| Ved Chirayeth | ARC | AIST-QRS-16-0004 | MiDAR-fused Supervised Machine Learning (SML) | Hines | 1/31/17 | Use of high resolution training sets to improve global scale moderate resolution data |
| Chris Mattman | JPL | AIST-QRS-16-0007 | Deep Web Search Analytics | Norton | 10/31/16 | Text analytics |
| John Readey | HDF Group | ACCESS15 -0031 | Object Store-based Data Service for Earth System Science | Hines | 5/31/17 | Foundation for data storage to improve access |
| Yehuda Bock | Scripps | AIST-QRS-16-0010 | Latency test of realtime warning systems in AMCE Cloud Computing | Quam | 9/30/17 | Feature Detection in near real time |
| Amy Braverman | JPL | AIST-QRS-16-0005 | Probablistic Climate Model Evaluation | Norton | 9/20/16 | What do you mean by similar? |
| Chris Lynnes | GSFC | AIST-QRS-16-0001 | Experiment with Data Containers in ESDIS Context | Little | 12/20/16 | Evaluation of alternative storage models for data to enable analysis |



AIST-16 Tasks-Machine Learning

| PI Name | Org | Prop # | Project Title | ESTO PM | Start Date | End Date | Machine Learning Elements |
|-----------------------------|------|--------|--|---------|------------|----------|---|
| Victor Pankratius | MIT | 48 | Computer Aided Discovery and Algorithmic Synthesis for Spatio-Temporal Phenomena in InSAR | Oza | 8/1/17 | 7/30/19 | Computer Aided Discovery |
| Jacqueline LeMoigne-Stewart | GSFC | 107 | TAT-C ML | Quam | 8/1/17 | 7/30/19 | Design space iterator |
| Branko Kosovic | NCAR | 79 | Fuel moisture content for improved fire prediction | Hines | 6/1/15 | 5/31/17 | Data assimilation |
| Andrew Michaelis | ARC | 137 | Framework for Mining and Analysis of Petabyte-size Time-series on the NASA Earth Exchange (NEX) | Oza | 8/1/15 | 7/31/17 | Time series analysis |
| Barton Forman | UMD | 24 | A mission planning tool for next generation remote sensing of snow | Quam | 9/1/17 | 8/30/19 | OSSE and assimilation |
| Ved Chirayath | ARC | 48 | NeMO-Net - The Neural Multi-Modal Observation & Training Network for Global Coral Reef Assessment | Oza | 9/1/17 | 8/31/19 | High resolution training sets improve moderate resolution imagery |
| Dara Entekhabi | GSFC | 49 | Autonomous Moisture Continuum Sensing Network | Hines | 9/1/17 | 8/31/19 | |
| Milton Halem | UMBC | 91 | An Assessment of Hybrid Quantum Annealing Approaches for Inferring and Assimilating Satellite Surface Flux Data into Global Land Surface Models. | Little | 6/1/15 | 8/31/19 | Assimilation, registration |
| Jeffrey Morisette | USGS | 124 | Advanced Phenological Information System | Hines | 9/1/17 | 8/31/19 | |
| Walter Jetz | Yale | 92 | Software workflows for remote sensing-based biodiversity change monitoring | Cole | 9/15/17 | 9/14/19 | |
| Jonathon Hobbs | JPL | 30 | Simulation-based Uncertainty Quantification | Norton | 10/1/17 | 9/30/19 | Statistics |
| Martyn Clark | NCAR | 81 | Climate risks in the water sector: Advancing the readiness of emerging technologies in climate downscaling and hydrologic modeling | Hines | 10/1/17 | 9/30/19 | |



Other AIST Activities Relevant to Machine Learning

- **AIST Virtual Reality Study**
 - Report Completed and undergoing revision
 - To be published by September 30, 2017
 - PoC: Shayna Skolnik (NAVTECA) and Emily Law (JPL)
- **AIST Block-chain Study**
 - Report completed 8/10/17 and undergoing revision
 - To be published by September 30, 2017
 - Interest in results by NRO, NSA, HEOMD (JSC, GRC), SCA N
 - PoC: Chris Dwan
- **Community Driven Workflow Recommendations**
 - CMAC11-0056 (funded by AIST) and AIST14-0032 (Seungwon Lee)
 - PoC: Jia Zhang, Carnegie Mellon University
- **AIST Machine Learning Workshop**
 - Fantastic Problems and Where to Find Them
 - Early CY18
 - Current PoC: Dan Crichton, JPL
- AIST-18 Solicitation (ROSES18)



Non-Government US Technology Investments

- <https://www.hpcwire.com/2017/08/17/google-releases-deeplearn-js-democratize-machine-learning/>
- <https://www.blog.google/topics/machine-learning/pair-people-ai-research-initiative/>

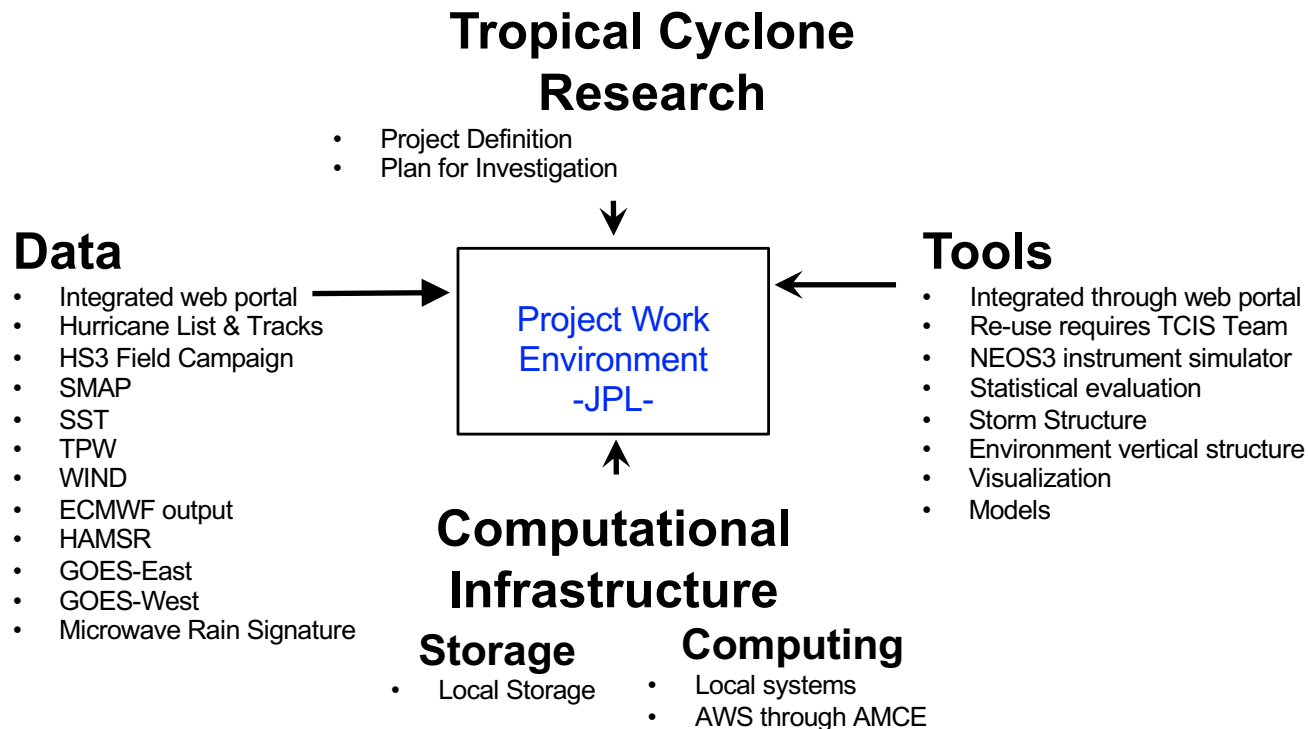


Candidate Useful ML Tools

- Implementations
 - TensorFlow
 - TensorFire (ML in the browser using jess)
 - arc-GIS server (Esri)
 - MathLab (Mathworks)
- Algorithms
 - Logistic regression
 - Symbolic regression
 - Random forest
 - Convolutional Neural Networks (CNN)
 - Deep Neural Networks (DNN)
 - Case-based reasoning
 - Restricted Boltzman Machines



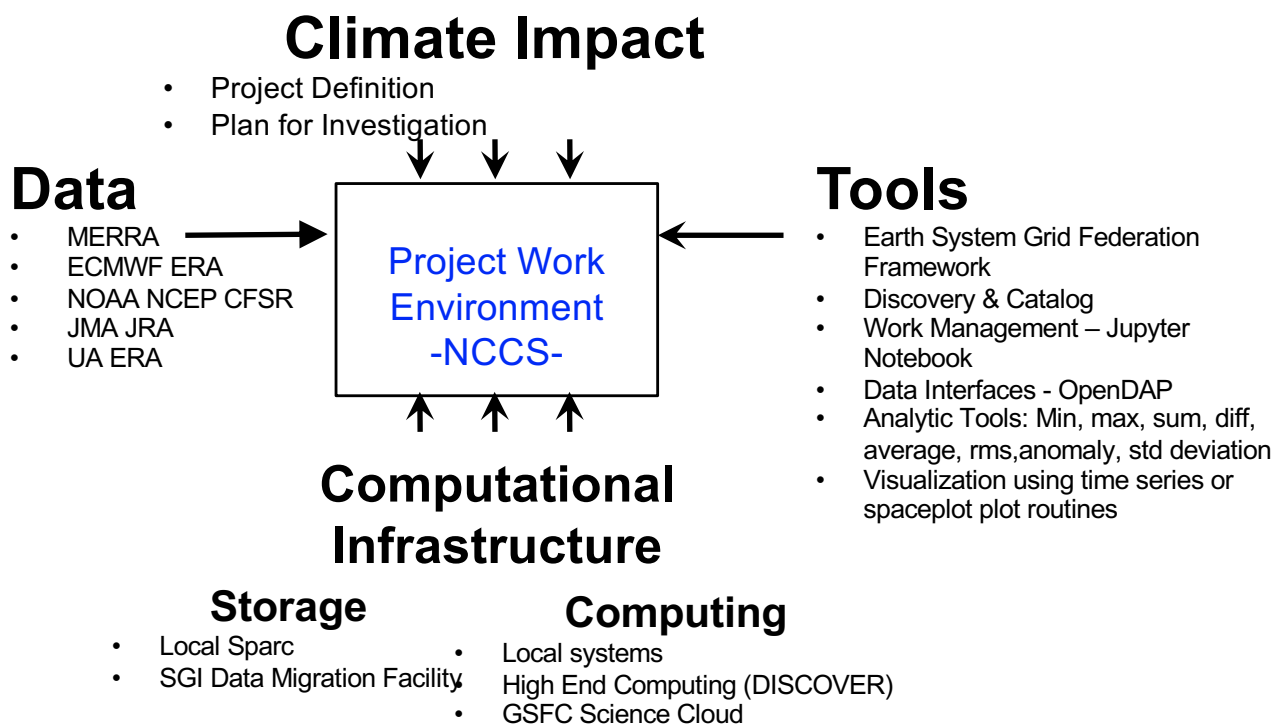
Tropical Cycle Information System (TCIS)



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<https://tropicalcyclone.jpl.nasa.gov/>



Earth Data Analytics System (EDAS) as an Analytic Center

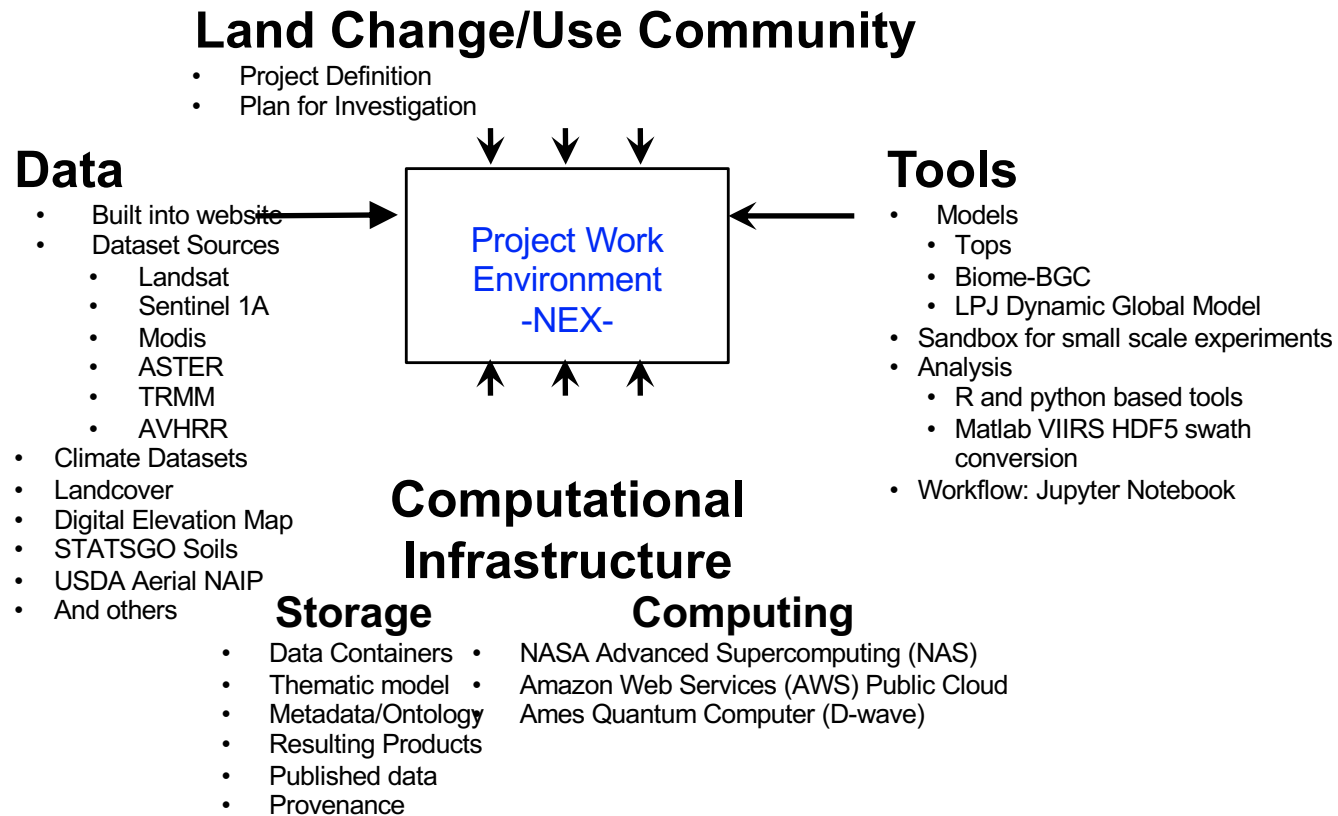


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<https://www.nccs.nasa.gov/services/Analytics>



NASA Earth Exchange (NEX) as an Analytic Center



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<https://nex.nasa.gov/nex>



Recent AIST Development Projects

- Data Wrangling
 - Data Container Studies reported at AGU2015, AGU2016
 - Ontology-based metadata (AIST14-0095)
 - Science Enterprise Service Bus (AIST14-0014)
 - In Situ-Satellite matchup service (AIST14-0086)
 - SciSpark (AIST14-0034)
 - Quantum Computing tools (AIST14-0096, AIST16-0091)
 - Multi-instrument data integration (AIST11-0077, AIST16-0048)
- Workflow tools (AIST16-0092)
- Analytic Tools
 - Regression analysis on Global Vegetation (AIST14-0115)
 - Pattern-based content analysis in GIS (AIST14-0027)
 - Uncertainty Quantification (AIST11-0011, AIST16-0030)
 - Computer-aided Discovery (AIST14-0036, AIST16-0048)
- Discipline Specific Tools
 - EcoSis Tools (AIST16-0118)
 - Hyperspectral image analysis tools (AIST16-0138)
- Modeling
 - Community re-organization biodiversity models (AIST16-0052)



Needs for Data Support Tools

- Interact with data from many sources
 - Identify known data sources
 - Determine data format characteristics
 - Add local, custom or authorized restricted data sources
 - Create interoperability interfaces
 - Stabilize data so it is persistent and available for others to evaluate
 - Refine existing metadata and add to it
 - Provenance
 - Reflect additional processing
 - Precision Ontology (Beth Huffer, AIST14-0095)



Needs for Analysis Tools

- Assume data is in a tractable format
- Perform what-if exercises on data analysis
- Track the analytic process performed
 - So it can be repeated and refined
 - Helps to accurately report results in publications
- Determine relevant past work and who did it
 - Published papers
 - Abstracts of research papers at AGU, AWS, ESIP, NTIS,
 - Abstracts of research awards in US Government



Needs for the Project Work Environment

- Human Factors Engineering
- A way to view the investigation that is comfortable to PI and the team
 - How do THEY think about the problem?
 - They must believe the AC reflects their view of the problem and will produce credible and defensible results
- Support seamless orchestration of tools operating on the data
 - It must feel natural in the context of the investigation
 - Accommodate What-if exercises at each stage in using different implementations of same algorithm or different algorithms for same function
- Permit PI to plan the resources needed for the Investigation
 - Assess the cost/performance tradeoff in planning
 - Unified Access (single sign-on)
 - Monitor consumption against the plan
- It must provide an automatic record of the work
 - Repeatability and credibility
 - Accelerate accurate details in publications



Needs for the Computational Resources

- Sufficient capacity
- Pay as you
- Special capabilities
- Affordability within the project budget
 - Advance prediction of the funding needs
 - Continuous update of plan vs. actual
- Security to preclude early release
- Different stages of the investigation invariably require different resources



Modeling in the Analytic Center

- Models capture and assess knowledge and understanding of the phenomena and processes
- Assess alternative models performance
- Propagate error estimates throughout



Analytic Center in a Box

- U-haul for a science investigation
- Create an AC for an Investigation
 - Low cost
 - Low latency
 - Define the expected needs for an investigation
 - Examine sensitivity of cost to investigation parameters
 - Provide standard tool selection
 - Permit easy incorporation of custom/unique tools



Analytic Center ConOps-Initialization

- Initialization
 1. Select environment framework and notebook
 2. Outline investigation strategy in Notebook
 - a. Identify, authorize and authenticate team
 3. Identify computing environment
 4. Select known data sources and interfaces
 5. Select known analytic tools
 6. Identify custom/local data sources
 - a. define interface
 7. Import custom or unique tools
 - a. Interface them into system
 8. Validation harmonization of all tools and data



Analytic Center ConOps-Analysis

- Seamless interaction with the data
 - Plug and play
- Tracking step-wise the analytic process
- TBD features
 - Analytic Center in a Box
 - Speed to set up
 - Ease of customizing tool entries
 - Ease of initializing data interfaces
 - Ease of initializing computing environment



Analytic Center ConOps-Publish

- Providing access to the data used upon which to base conclusions has become a necessity for publications to be accepted
- New NASA requirements for publications make distribution to all the mandatory delivery points time consuming
- AGU and the majority of the publishing industry are working on a template for submission of articles for peer review
 - Extensive supporting documentation
- The publication tool set should enable meeting these new and old bureaucratic requirements
 - Collect artifacts as the Investigation proceeds so they can be assimilated into the publication submission
 - Crediting references and data sources
 - Enabling the access to the data for validation
 - Distribution of the documents to the required distribution points
 - Configuration management
 - Reviewer comment resolution and tracking



AIST MCS End State – PI Perspective

- Relationship with AMCE is defined by a Service Level Agreement
 - PI will work with AMCE Cloud Project Manager and Cloud System Administrator to estimate their consumption of cloud resources
 - Agreement to be monitored for computer security and accounting purposes
- Each PI has access to the full range of AWS Cloud Services
 - Choices of what type and how much are left to the PI and his team
 - As new services are created by AWS, they are made available to PI as soon as possible
 - PI and his team are familiar with the services they need and how to use them in the AIST MCS
- AIST Cloud Orientation Class for PI and entire Team
 - Outline rules of engagement: consumption and usage, fiscal restraints and computer security measures
 - Emphasis on value of relationship, as defined by an agreement among all parties, not on regulatory compliance
- Reports are pushed to PI describing consumption, and logins
 - PI's are expected to review these and work with AMCE Cloud PM/SA to resolve any issues
- Financial Management
 - During early phase startup, AIST PM funds AWS and allocates funding to each Project
 - Once orientation/debugging is complete, the PI pays for computing services as he uses them, using money from his project
- AIST MCS supports both NASA Govt and Contractor employees, non-US citizens, University Researchers and commercial companies
 - PI and a lead System Administrator are Privileged Users
 - Researchers are Users



Cloud Service Roles Defined

- **Billing Manager** is the cloud services team member responsible for the creation of as well as financial monitoring and oversight of all AWS linked billing accounts. (SGT AWS Account, eventually SEWP Master)
- **Master Cloud Administrator** is the cloud services team member responsible for Identity and Access Management (IAM) and technical administration and support for the AWS linked billing accounts. (SGT AWS Account, eventually ARC SEWP manager responsibility)
- **Account Coordinator** is the person within the organization responsible for requesting/establishing, funding, and first line oversight of an IAM account. (AIST Funder Role)
- **CyberSecurity Administrator** is the person responsible for the Identity and Access Management within a cloud account, establishing security policies, and establishing a separate cloud instance as a log server to log, monitor, and review all cloud instance(s) activity under their cloud account in compliance with LaRC IT Security requirements. This role may be fulfilled by either a cloud services team member or a person designated by the Account Coordinator. (AIST Role possibly combined with Cloud SA person, but separate role)
- **Cloud Systems Administrator (SA)** is the person responsible for creating cloud instances, technical administration and support, specified tagging and establishing automated billing alerts, and performing emergency stops of an instance of the authorized AWS IAM accounts. This role may be fulfilled by either cloud services team member or a person designated by the Account Coordinator. (AIST Role)
- **Cloud Project Manager** is the person within the organization responsible for providing funding to the Account Coordinator for the project and monitoring the project's usage. This is an optional role that may be designated by the Account Coordinator. (AIST Role)
- **Privileged User** is the person within the organization responsible for starting/stopping and provisioning cloud instances. This role may be fulfilled by either cloud services team member or a person designated by the Account Coordinator. (PI System Admin)
- **End User** means any individual or entity that directly or indirectly accesses or uses a cloud instance and/or cloud services under Langley Services. This person has the ability to stop cloud instances. (Project Team)
- **Technical Monitor** is the Technical Monitor (TM) for the Contractor Task Order providing cloud services. (Government procurement activity, initial LaRC Fran Risinger, then AIST, then SEWP)